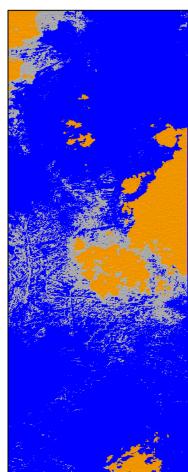




## EO10nboard Cloud Cover Detection Validation Preliminary Report



March 11, 2003

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#### Team Members



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#### Agenda



- ♦ Introduction Dan Mandl & Jerry Miller
- ♦ Requirements Dan Mandl
- ◆ Development Effort Bruce Trout
- ◆ Cloud Assessment Procedure Michael Griffin
- ◆ Conclusion Jerry Miller & Dan Mandl



# Intro: Cloud Cover Assessment Concept



- Rationale: On board cloud assessment has the potential to considerably reduce the resources on downlink for unwanted scenes.
- Concept: Flight validate an onboard cloud cover detection algorithm and determine the performance that is achieved on the Mongoose V
- Approach:
  - Formulate and test a cloud cover determination algorithm that is compatible with Hyperion sensor measurements
  - Using MIT / LL provided algorithm, implement and test code to execute on EO-1 platform
  - Uplink and execute code updates onboard EO-1, and evaluate its performance on orbit
- $\bullet$  TRL In = 5

TRL Out = 6



#### Intro: Initial Results



- Final onboard cloud cover assessment of an EO-1 8 second (.75 Gbyte) Hyperion scene was expected to take hours but instead took less than 30 minutes
- Streamlined algorithm by:
  - Performing level 0 on all data and then selecting the needed 6 bands
  - Converted level 0 data to radiance (level 1R) one scan line (256 pixels) at a time
  - Performed pixel by pixel cloud assessment
- Can perform onboard cloud assessment faster with the following capabilities:
  - Subsampling of raw data (can get close to same results without processing all data)
  - User defined area of interest within image and only process that portion
  - Direct access to science recorder
  - Cloud assessment algorithm can be expanded since we had more margin than expected
- For 20 test cases on ground, performed cloud assessment within 5% for major test cases



#### Intro: Comparison of ESTO Onboard Cloud Cover Studies



	1999	2002/2003
Test type	Simulation	On-orbit
Instrument	NOAA14 AVHRR multispectral	EO-1 Hyperion hyperspectral
Bands	.5868725-1.10-3.55- 3.93-10.3-11.3-11.5-12.5	.556686-1.25-1.38-1.65
Processing scenario	Real time	About 1 orbit
Processor	Commercial Power PC 750, 233 Mhz, 450 MIPS	Rad hard Mongoose V, 12 Mhz, 6-7 MIPS
Operating System	Linux	VxWorks
Software Preprocessing	Albedo, radiance and brightness test	Raw data>> L0 >>Level 1b >> reflectance
Tetsed algorithms	Land, sea, day, night, clouds, ice, snow, sand, sun glint	Differentiate clouds from ice, snow, sand and water



#### Intro: Spacecraft



#### **♦** Two primary Science Instruments

- Advanced Land Imager
  - 10m resolution
  - Visible imager
  - Questionable if can access data onboard due to onboard format

#### Hyperion

- 30m resolution
- Hyper spectral imager (220 bands)
- Data access onboard for cloud detection

#### Orbit

- 705 km Altitude
- ~15 day Repeat track
- 98.7 degree inclination

# Earth Observing-1

## Intro: EO-1 Extended Mission Testbed Activities



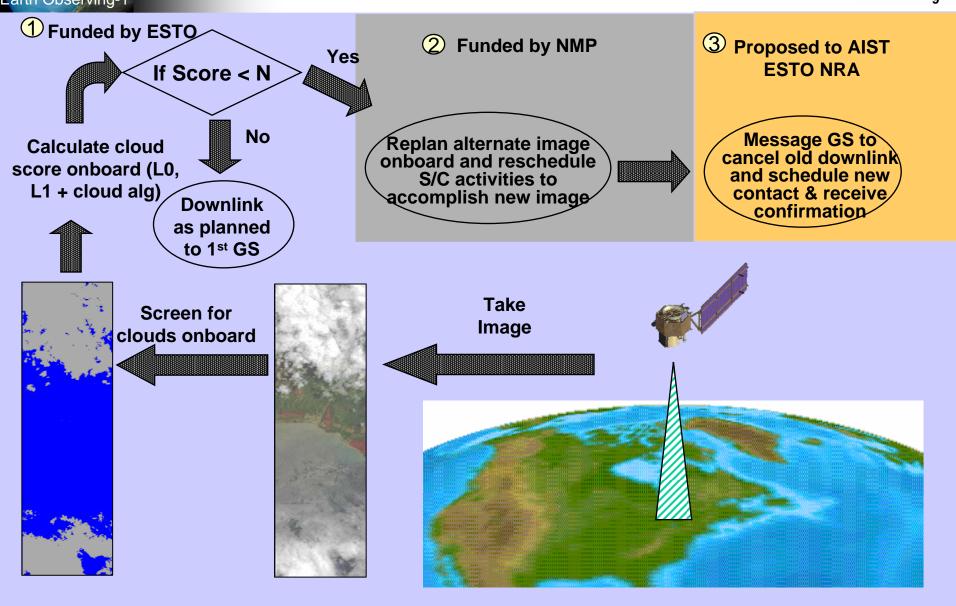
Operational End-to-End Autonomous Onboard **Testbed** Coordination **Communications Processing Autonomous Science** Experiment(ASE) Onboard Cloud Cover Preliminary EO-1 ♦ Migration of ST6 onto **Detection Validation Autonomy Experiment** EO-1 \$180K Onboard planning \$ TBS Onboard feature detection (5) EO-1, Terra, Aqua ♦ Dynamic SW Bus Sensorweb Demo \$720K ◆ Uses MODIS inst center Sensorweb Simulator to detect volcanoes ◆ Related activity but ◆ Uses ASE to coord image not spawned by EO-1 collect autonomously ♦ M. Seablom no additional funding needed \$200K 6 Smart Antenna ♦ Ground phased array EO-1/ Gnd Sensor Sensorweb ♦ Cell tower com to sat ♦ Sensors in Huntington Hyperspectral AIST ESTO NRA Proposal Botanical Garden trigger Compression WG (3) Dynamic Resource (4) EO-1 image Onboard data mining Management no additional funding needed ♦ Onboard intelligent **Autonomous** image compression scheduling of Ground Intelligent Distributed Working group Station by satellite Spacecraft Technology AIST ESTO NRA Proposal Testbed: NMP (JPL) Related activity Funded by ESTO Funded by NMP Proposed activity

Note: Numbered boxes are detailed in following slides.



## Intro: EO-1 Onboard Cloud Cover Detection With Onboard Replanning







#### Intro: Related Ongoing Feature Detection Efforts

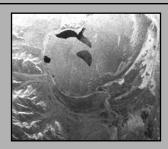


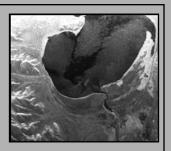


#### **Funded by NMP**

#### **Autonomous Change Detection**

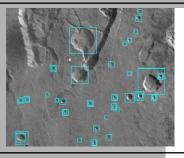
- · Ice formation/retreat, flooding
- Atmospheric Change
- Volcanic processes (Lava, mud, plume)





#### **Autonomous Feature Identification**

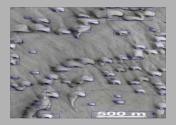
- Volcanic cinder cones and craters
- Impact craters
- Sand dunes

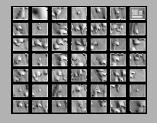




#### **Autonomous Discovery**

Identify features which differ from the background





- Downlink science products: science events, features not raw data
- Achieves 2x-100's x data reduction!



#### Cloud Cover Assessment Top Level Requirements



- Implement, test, and upload WARP flight software update to perform cloud cover processing on Hyperion SWIR / VNIR image files as requested
- Extract pixel read out values from these files for bands designated for cloud cover assessment use. (Includes both SWIR / VNIR bands)
- Perform radiometric calibration to Level
- Perform cloud cover assessment and telemeter results to the ground
- Provide mechanisms to control cloud cover processing and provide reporting of cloud cover processing status



# Level 0 and Level 1Processing Requirements



- Perform playback of requested SWIR / VNIR image data files stored on WARP
- Synchronize on header for 1<sup>st</sup> science data packet
- ◆ Extract each spatial pixel read out value from this packet for bands designated for cloud cover assessment use
  - VNIR bands 0.55 (band 20), 0.66 (band 31), 0.86 (band 51)
  - SWIR bands 1.25 (band 110), 1.38 (band 123), 1.65 (band 150)
  - Read out value extraction involves stripping 12 least significant bits of 2 byte value
- ◆ Apply level 1 calibration to each level 0 data sample



#### Cloud Cover Detection Requirements



١,

- Perform pixel by pixel testing using reflectance data to determine which pixels are cloud covered.
- Cloud coverage for a given pixel will be determined based on results of a series of tests as described in the MIT presentation. Types of tests will include:
  - Reflectance Threshold tests. Tests reflectance value for a given spectral pixel relative to a predefined threshold.
  - Ratio test. Tests ratio of reflectance values for 2 different bands for a given pixel relative to a predefined threshold.
  - Normalized Difference Snow Index (NDSI) test. Tests differences of 2 bands divided by sum of the 2 bands relative to a predefined threshold value.
  - Combo test. Uses results of NDSI and Reflectance threshold tests.
- Statistics to be provided which provide total tested and cloudy pixels, and percentage cloudy.

#### Dev Effort – SW Environment

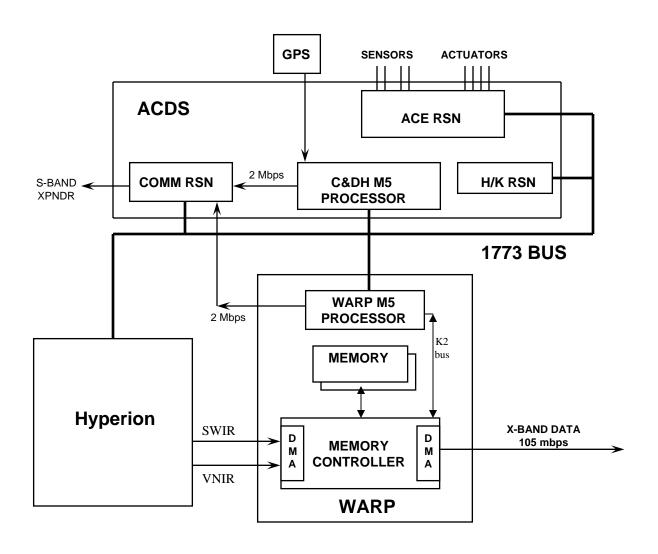


- ◆ Two Mongoose 5 (M5) Processors:
  - C&DH, WARP
  - 12MHz, ~6 MIPS, 256 MB RAM on each M5
- ♦ Both M5's running VxWorks 5.3.1
- ♦ WARP M5 unused except for collection, S-band downlink events
- WARP M5 has access to spacecraft bus for telemetry, commanding



## Dev Effort: EO-1 Data Architecture

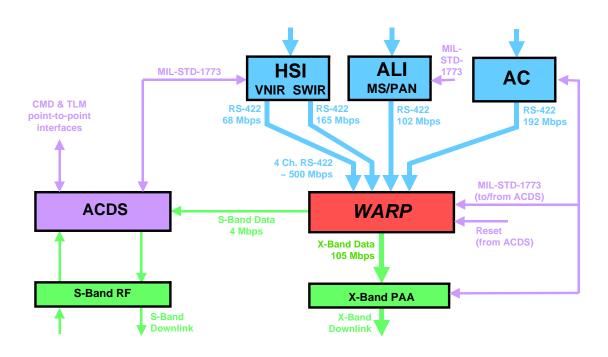






## Dev Effort: WARP Data Flow



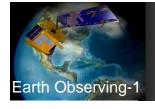




## Dev Effort: WARP Block Diagram



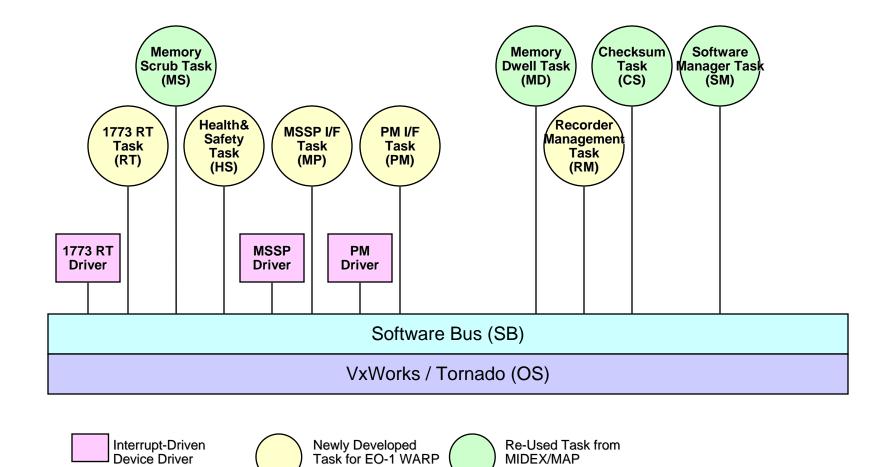
Science Data Input CMD/TLM & Processing S-Band Data 4 Mbps 1773 RS-422 Downlink Output Bulk Data Storage X-Band I & Q **WARP FODB** RS-422 RF Memory Memory Memory RSN Input Input Exciter Interface Board Board Board Board Board 24 Gbits 24 Gbits



Device Driver

## Dev Effort: Existing WARP FSW Architecture





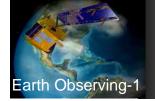
MIDEX/MAP



## Dev Effort: Cloud Cover Patch Concept



- Will run as part of memory dwell, when not dwelling MD currently does nothing except wait for messages on the software bus
- Memory Dwell is lowest priority task except idle.
- S-Band playback control flow messages will be re-routed to and from the MP task to the MD task by patching the software bus routing tables.
- CC Code will run whenever data ready message is sent from RM
- MD will utilize all spare CPU in system
- Health and safety CPU hogging check will be patched out with NOPs



## Dev Effort: Cloud Cover Detection Software Memory Usage



## ◆ Cloud cover SW patches fit between tasks (Gaps ~100kBytes)

Data Value	Data Type	Size
Warp File IDs	ground uplink	8 bytes
Solar Zenith Angle & Julian Day	ground uplink	24 bytes
Dark Noise Offset values (6 bands * 256 pixels * 2 bytes)	calculated	~3 Kbytes
Calibration factors (6 bands * 256 pixels)	stored	~6 Kbytes
Solar flux values (6 bands)	stored	24 bytes
S-Band data local buffer	recorded data	~929k
		(existing buffer)
Image level 0 data	generated	8 Mbytes
		(for 12s image)
Cloud Cover Test Thresholds (<10)	stored	40 bytes
Cloud Cover Statistics and Telemetry	generated	< 1k bytes



## Dev Effort: SW Test Approach



- ◆ Algorithms were integrated and tested first on a PC based simulation system using files for input test data 9/02
- ♦ Patch test 8/02
  - Prove that we can patch the WARP Mongoose V without a full fidelity test bed
  - Patched No-op
- ◆ Level 0 bandstripping test 11/02, 12/02 and 1/03
  - Test of full kernel load needed for later loading of CASPER
  - Test capture of playback data from WARP to Mongoose
  - Test level 0 bandstripping of data
- ◆ Level 1 and onboard cloud assessment 3/10/03
  - Test conversion to level 1
  - Test cloud algorithm
  - Measure performance



## Dev Effort: Development Challenges



- WARP test bed limitations
  - WARP Wide Band Recorder and associated interfaces do not exist
  - WARP M5 Available Memory limited to 32 Mbytes versus the onboard memory which has 256 Mbytes
- ◆ Revised load process and checksum process

#### Cloud Cover Estimation Procedure



◆ From calibrated Hyperion radiance data, convert to top-of atmosphere (TOA) reflectance and estimate on a pixel-by-pixel basis the extent of cloud cover in a scene.

1. Convert radiance data to TOA reflectance

Use pre-computed band solar flux values, earth-sun distance ratio, and the solar zenith angle

2. Process each frame (or line) of data

Determine which pixels are cloud-covered

Distinguish land, water, snow or ice from clouds

3. Produce cloud cover statistics for the scene



#### 1. Radiance to TOA Reflectance



#### - Procedure -

- Obtain calibrated level 1B radiance data
  - Large part of cloud cover effort is focused on this task
  - 1 frame (256 samples by 6 bands) at a time
- Obtain from telemetry or other means for the Hyperion scene
  - Earth-sun distance ratio d<sub>e-s</sub>
  - Cosine of the solar zenith angle  $\mu_0$
  - Band Solar Flux values S<sub>0,i</sub>
- For each band i use the following formula to convert the calibrated Hyperion radiance L<sub>i</sub> to reflectance ρ<sub>i</sub>
- ◆ Final product is one TOA reflectance value for each band at each pixel
  - $\rho$ (256,6) for a single Hyperion frame

$$ho_i = \left[ rac{\pi}{\mu_0 \, S_{0,i} / d_{e-s}^2} \right] L_i$$





- Basic Tests -

The cloud cover algorithm uses only 6 bands of Hyperion data

- 0.56, 0.66, 0.86, 1.25, 1.38, 1.65  $\mu$ m

0.56  $\mu$ m : used w/ 1.65  $\mu$ m to compute the snow index

0.66 μm : basic cloud reflectance test channel

0.86  $\mu$ m : used w/ 0.66  $\mu$ m in NDVI-like ratio test

1.25  $\mu$ m : desert/sand discrimination

1.38  $\mu$ m : high cloud test channel

1.65  $\mu$ m : used w/ 0.56  $\mu$ m to compute the snow index

♦ On-board processing limitations requires small number of bands

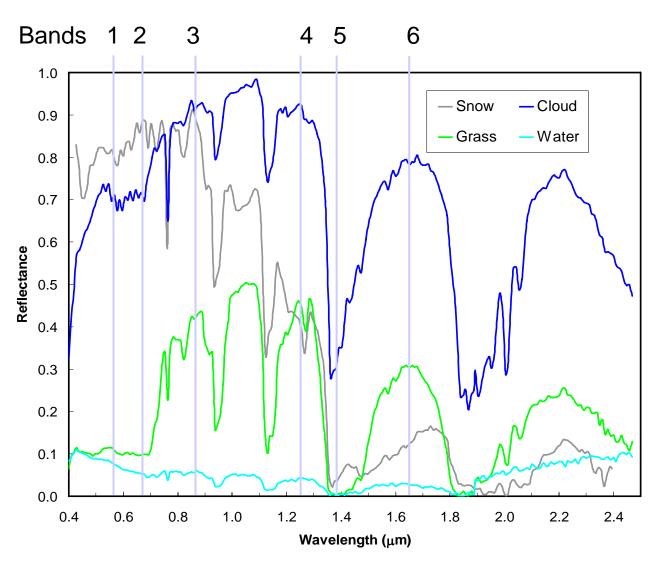
Each test utilizes TOA reflectance data

◆ 20 Hyperion scenes of varying surface and cloud features were used to define test thresholds



# Spectral Band Locations With Sample Reflectance Curves







## Cloud Cover Detection Algorithm



**Start** 6 channels used 0.56, 0.66, 0.86, 1.25, 1.38, 1.65 μm  $\frac{\rho_{.66}}{\rho_{.66}} < \rho_{T_3}$  $\rho_{1.38} > \rho_{T_1}$ **High/Mid Cloud**  $T_5 > NDSI \ge T_6$  $ho_{.86}$ Ν  $NDSI = \frac{\rho_{.56} - \rho_{1.65}}{\rho_{.56}}$  $\rho_{.56} + \rho_{1.65}$ **Vegetation** Ν  $DSI = \frac{\rho_{1.25} - \rho_{1.65}}{\rho_{1.65}}$  $\rho_{.66} \ge \rho_{T_7}$ **Bare Land**  $\rho_{1.25} + \rho_{1.65}$ Water Υ  $\rho_{1.38} < \rho_{T_1}$  $\frac{\rho_{.66}}{\rho_{T_3}} \ge \rho_{T_3}$ Ν Low/Mid Cloud **Vegetation**  $ho_{.86}$  $\rho_{1.25} > \rho_{T_A}$ Snow / Ice Υ Ν  $DSI > T_5$ **Desert / Sand** 5a  $NDSI < T_7$ N  $T_6 > \text{NDSI} \ge T_7$  $NDSI > T_8$  $T_5 - T_8$  are  $\rho_{T1} - \rho_{T4}$  are reflectance thresholds index thresholds

NDSI: Normalized Difference Snow Index, DSI: Desert/Sand Index



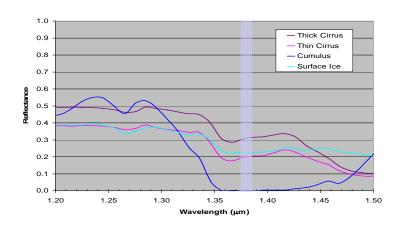
# Cloud Cover AlgorithmNIR Absorption Band Tests -



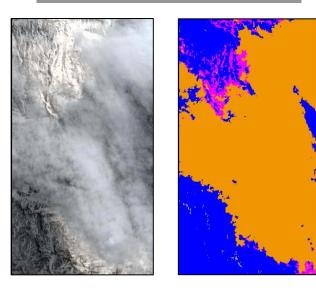
Test 1: High/mid cloud reflectance threshold

$$\rho_{1.38 \, \mu m} > \sim 0.1$$

- Only high clouds are typically observed in this channel
- Strong water vapor absorption masks most low level/surface features
- Under dry conditions, surface features such as ice and snow can be observed and mistaken for clouds
- Further vegetation and snow/ice discrimination tests are necessary to isolate clouds



#### **Cheyenne Wyoming**



Cloud-free, Low/Mid cloud, Mid/High cloud





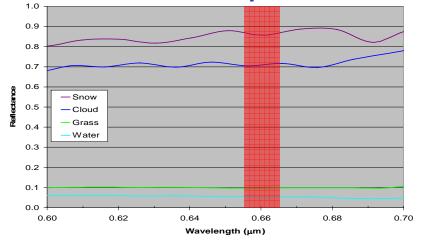




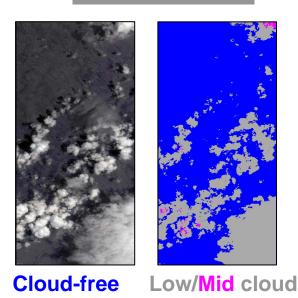
#### Test 2 : Red channel reflectance threshold

 $\rho_{0.66 \, \mu m} > \sim 0.3$ 

- Assumes low reflectance of most vegetation, soil and water surfaces in the red region of the spectrum
- Snow, Ice, bright desert/sand surfaces and clouds should pass this test



#### Kokee Hawaii



No (N)	Yes (Y)
Vegetation	Snow / Ice
Soil	Desert / Sand
Water	Some Vegetation
	Clouds





#### - Visible/NIR Ratio Test -

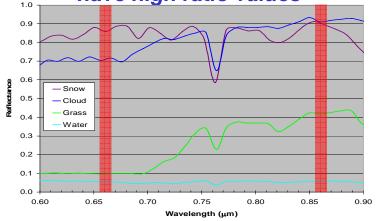


#### Test 3 : VIS/NIR ratio test

$$\rho_{0.66 \, \mu m} / \rho_{0.86 \, \mu m} > \sim 0.7$$

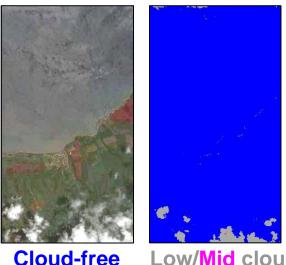
- Discriminates vegetative surfaces whose reflectance varies strongly from Visible to NIR
- Vegetative and soil surfaces exhibit small ratio values.

Clouds, desert/sand, snow and ice surfaces have high ratio values



No (N)	Yes (Y)
Vegetation	Snow / Ice
	Desert / Sand
	Clouds





Low/Mid cloud





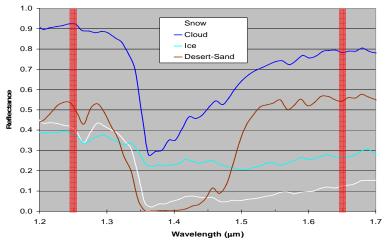


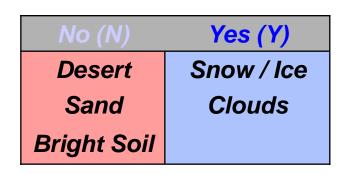


#### Test 4 : Desert Sand Index (DSI)

$$DSI = \frac{\rho_{1.25} - \rho_{1.65}}{\rho_{1.25} + \rho_{1.65}} > -0.01$$

- Discriminates bright soil and sand surfaces whose reflectance increases slightly from 1.25 to 1.65 μm
- Clouds, snow and ice reflectance tends to decrease over this range





#### Suez Canal



**Cloud-free** 





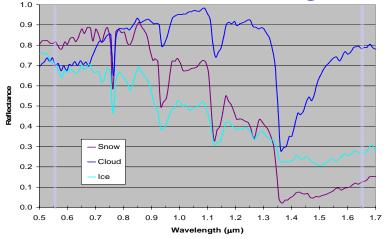
## Cloud Cover Algorithm - SWIR Snow/ice/cloud Test -

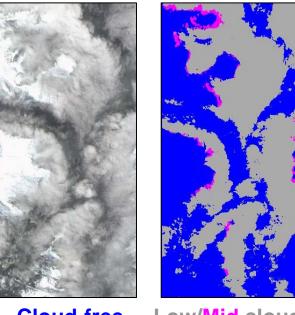


#### Test 5 : Normalized Difference Snow Index (NDSI)

$$NDSI = \frac{\rho_{0.56\,\mu\text{m}} - \rho_{1.65\,\mu\text{m}}}{\rho_{0.56\,\mu\text{m}} + \rho_{1.65\,\mu\text{m}}}$$

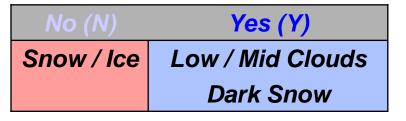
- Some sparse or shadowed snow (in mountains) can pass test
- Cloud-free snow generally displays





**Cloud-free** 

Low/Mid cloud





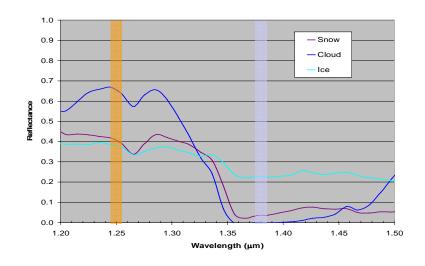


#### - SWIR Reflectance Tests -



#### SWIR Reflectance Tests

- Test 6  $\rho_{1.25 \mu m} > \sim 0.35$
- *Test* 7  $ρ_{1.38 \mu m} < ~ 0.1$
- Eliminates most snow/ice
- Low/Mid clouds should pass tests



#### **Bering Sea**



**Cloud-free** 

Low/Mid cloud





# Cloud Cover AlgorithmTest Case Results -



- The following slides show results from the cloud cover algorithm for a selection of Hyperion scenes
- One or two segments (1000 lines each) of the overall Hyperion scene are displayed
- Cloud cover estimates (percent of displayed scene covered by all clouds) is shown at the bottom
- ◆ Examples are meant to highlight successes and failures of algorithm

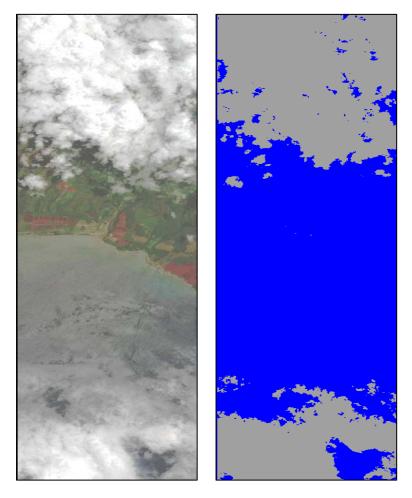
<u>Colors</u>

**Cloud-free** 

Low/Mid cloud Mid/High cloud



Kokee Hawaiji Lines 3200 - 4200

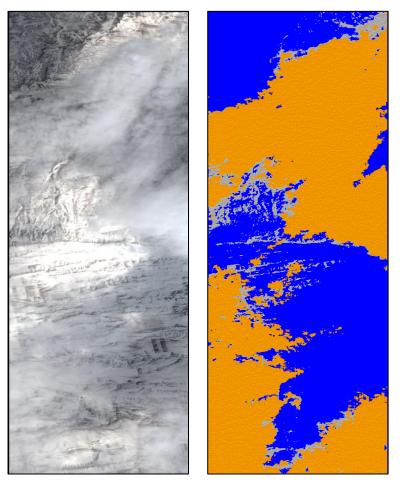


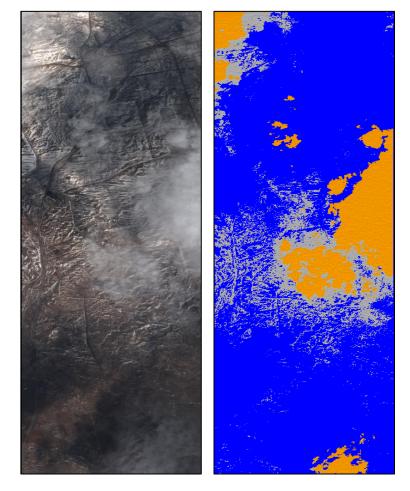
Total Cloud: 41.3 %

Total Cloud: 6.8 %

Success	Discriminates land/cloud, land/water
Failure	Misses some darker cloud over water

## Lines 500 - 1500 heyenne Wyomin Lines 2000 - 3000





Total Cloud: 58.9 %

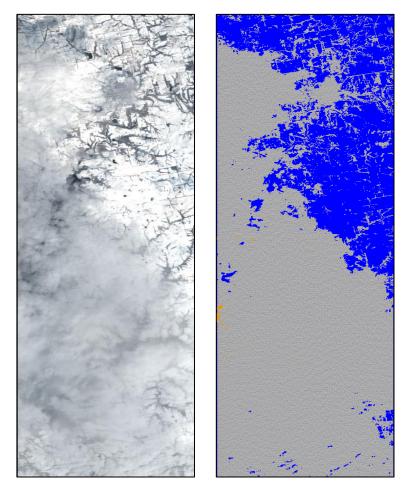
Total Cloud: 27.0 %

Success	Snow/cloud, ice cloud
Failure	Difficulty with shadowed snow cover



**Lines 0 - 1000** 

Lines 2100 - 3100



**Total Cloud: 72.6 %** 

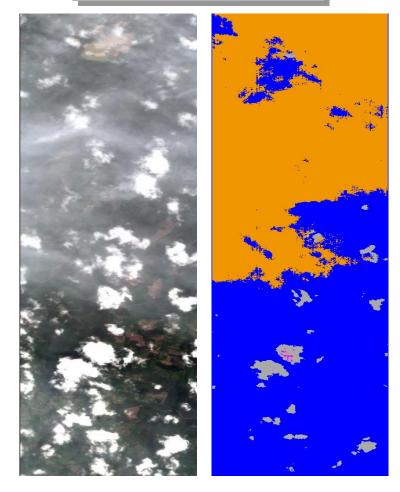
Total Cloud: 18.6 %

Success	Bright Snow/cloud discrimination
Failure	Some snow cover flagged as cloud

#### Chiefs Island

#### **Total Cloud: 68.9 %**

#### Lake Pontchartrain

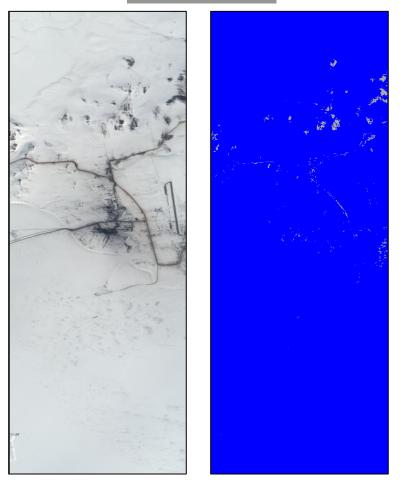


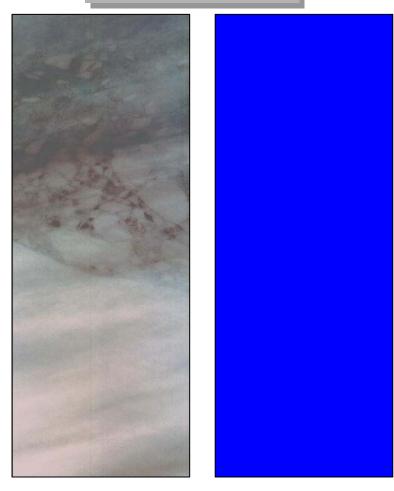
Total Cloud: 48.6 %

Success	Detects Cirrus, Cumulus
Failure	Cloud Cover underestimated

#### Bering Sea

#### Larsen Ice Shelf





Total Cloud: 0.7 %

Total Cloud: 0.0 %

Success	Bright Ice, snow all flagged clear
Failure	Small amount of dark snow features

# Suez Canal Chernobyl

Total Cloud: 0.3 % Total Cloud: 0.0 %

Success	Bright sand, soil all flagged clear
Failure	Small amount of bright soil



# Summary of Cloud Cover Algorithm Performance



#### Schedule calls for first on-board test in December 2002

- Algorithm results are encouraging
- On-board cloud cover detection accuracy requirements are not stringent (10-15 %)
  - Only need to know if scene is clear enough for user
  - Simple algorithms with limited # of bands sufficient
- Algorithm does a good job not classifying bright surface features (snow, ice, sand) as clouds
- ◆ Difficulties with dark snow and dark/shadowed features
  - Adjustment of thresholds (e.g., geographical, seasonal) may improve results
- ♦ Areas for future enhancements/improvements
  - More sophisticated algorithms
  - More bands
  - More precise validation of actual cloud cover



#### Conclusion



- Discovered many methods to streamline onboard cloud assessment
- ◆ Big driver to onboard cloud assessment is precision required
  - For many applications, accuracy within a 5% is adequate thereby allowing shortcuts